

The Asymptotical Error of Broadcast Gossip Averaging Algorithms

Fabio Fagnani **Paolo Frasca**

Dipartimento di Matematica

Politecnico di Torino, Italy



IFAC World Congress
Milan, September 1, 2011

Averaging in networks: Challenges

Distributed averaging is a building block to solve estimation problems in sensor and control networks.

Depending on the application, in distributed averaging we need to

- design efficient algorithms with little communication requirements
- analyze their performance
 - in terms of both speed and **accuracy**
 - as a function of the network topology and size (large networks)

Averaging in networks: Formal problem statement

Set-up:

- a set of nodes \mathcal{V} of cardinality N
- a graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ (undirected in this talk)
- data at the nodes: $y_v \in \mathbb{R}$ for all $v \in \mathcal{V}$

Goal: estimate the average $y_{\text{ave}} = N^{-1} \sum_{v \in \mathcal{V}} y_v$.

Constraint:

- avoid synchronous node updates
- use instead directional asynchronous communication

→ use randomized broadcast communication

Averaging in networks: Formal problem statement

Set-up:

- a set of nodes \mathcal{V} of cardinality N
- a graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ (undirected in this talk)
- data at the nodes: $y_v \in \mathbb{R}$ for all $v \in \mathcal{V}$

Goal: estimate the average $y_{\text{ave}} = N^{-1} \sum_{v \in \mathcal{V}} y_v$.

Constraint:

- avoid synchronous node updates
- use instead directional asynchronous communication

→ use randomized broadcast communication

Broadcasting Gossip Algorithm (BGA): definition

Broadcast Gossip Algorithm

```

1: for  $v \in \mathcal{V}$  do
2:    $x_v(0) = y_v$ 
3: end for
4: for  $t \in \mathbb{Z}_{\geq 0}$  do
5:   Sample node  $v$  from a uniform distribution over  $\mathcal{V}$ 
6:   for  $u \in \mathcal{V}$  do
7:     if  $u \in \mathcal{N}_v$  then
8:        $x_u(t+1) = (1-q)x_u(t) + qx_v(t)$ 
9:     else
10:       $x_u(t+1) = x_u(t)$ 
11:    end if
12:  end for
13: end for

```

Mixing parameter:
 $q \in (0, 1)$

A. G. Dimakis, S. Kar, J. M. F. Moura, M. G. Rabbat, and A. Scaglione. Gossip algorithms for distributed signal processing. *Proceedings of the IEEE*, 98(11):1847–1864, 2010

Preliminary results & definitions

Proposition (Convergence)

If \mathcal{G} is connected, then there exists a random variable x^* such that almost surely $\lim_{t \rightarrow +\infty} x(t) = x^* \mathbf{1}$.

Proposition (Martingale property)

Let $x_{\text{ave}}(t) = N^{-1} \sum_{v \in \mathcal{V}} x_v(t)$.

Then $\{x_{\text{ave}}(t)\}_t$ is a martingale (w.r.t. $x(t)$) and $\mathbb{E}[x^*] = x_{\text{ave}}(0)$.

However, x^* is not equal to $x_{\text{ave}}(0)$. The goal of this work is studying

$$\beta(t) = |x_{\text{ave}}(t) - x_{\text{ave}}(0)|^2,$$

and in particular the limit $\mathbb{E}[\beta(\infty)] := \lim_{t \rightarrow \infty} \mathbb{E}[\beta(t)]$

Preliminary results & definitions

Proposition (Convergence)

If \mathcal{G} is connected, then there exists a random variable x^* such that almost surely $\lim_{t \rightarrow +\infty} x(t) = x^* \mathbf{1}$.

Proposition (Martingale property)

Let $x_{\text{ave}}(t) = N^{-1} \sum_{v \in \mathcal{V}} x_v(t)$.

Then $\{x_{\text{ave}}(t)\}_t$ is a martingale (w.r.t. $x(t)$) and $\mathbb{E}[x^*] = x_{\text{ave}}(0)$.

However, x^* is not equal to $x_{\text{ave}}(0)$. The goal of this work is studying

$$\beta(t) = |x_{\text{ave}}(t) - x_{\text{ave}}(0)|^2,$$

and in particular the limit $\mathbb{E}[\beta(\infty)] := \lim_{t \rightarrow \infty} \mathbb{E}[\beta(t)]$

Main result

Bounding the error introduced at each time step as

$$|x_{\text{ave}}(t+1) - x_{\text{ave}}(t)| \leq q \frac{d_{\max}}{N} L, \quad \text{where } L \geq \max_{u,v} |x_u(0) - x_v(0)|,$$

and exploiting the martingale property, we can prove

Theorem (Uniform Mean Square Error Bound)

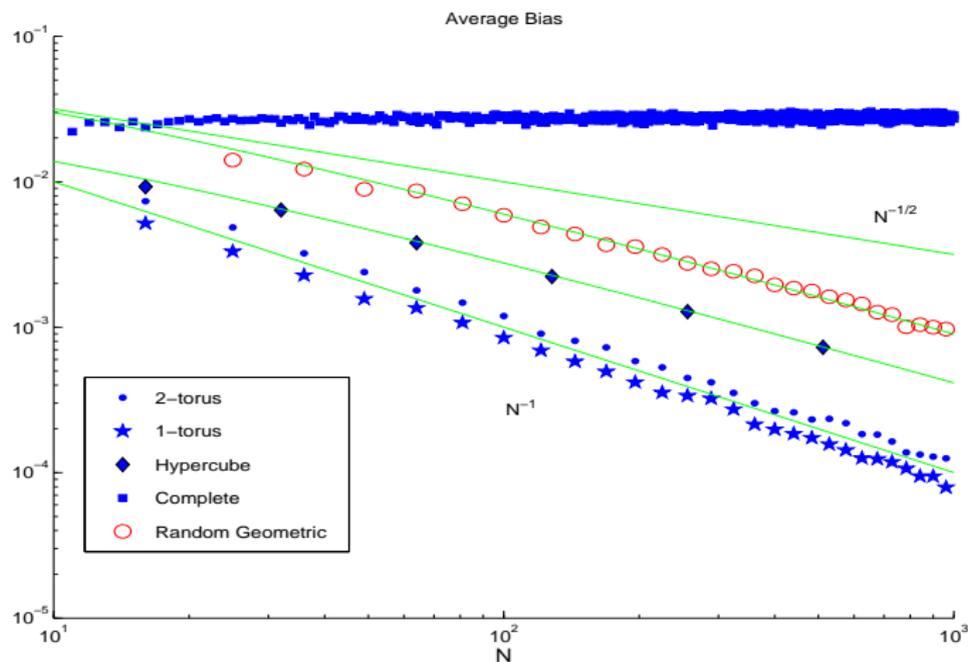
Let \mathcal{G} be connected, λ_1 be its spectral gap and d_{\max} be the maximum degree of its nodes. Then,

$$\mathbb{E} \left[\sup_{t \in \mathbb{N}} \beta(t) \right] \leq 8 L^2 \frac{q}{1-q} \frac{d_{\max}^2}{N \lambda_1}$$

Think of large networks....

Simulations: $\beta(\infty)$ vs N

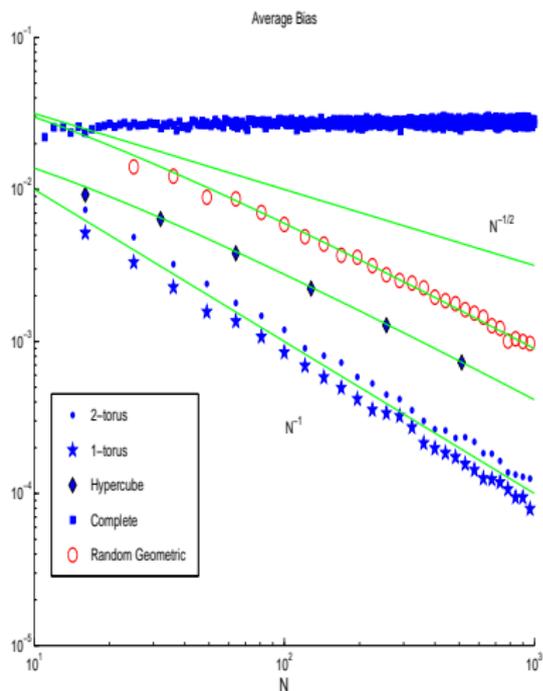
Varying size in example sequences



Solid lines are proportional to $N^{-1/2}$, $\log N/N$, $\log N/N$ and N^{-1}

Simulations: $\beta(\infty)$ vs N

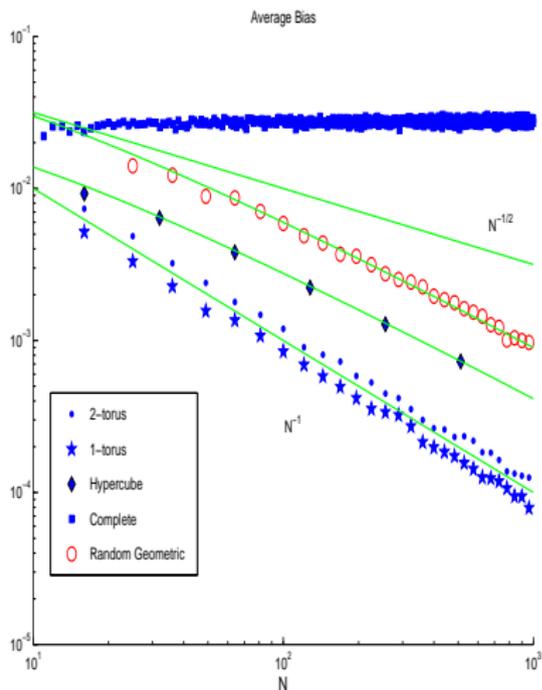
Varying size in example sequences



Solid lines are proportional to $N^{-1/2}$, $\log N/N$, $\log N/N$ and N^{-1}

Simulations: $\beta(\infty)$ vs N

Varying size in example sequences



graph	degree	$\beta(\infty)$
complete	$N - 1$	constant
ring (1-torus)	2	$1/N$
2-torus	4	$\sim 1/N$
hypercube	$\log N$	$\sim \frac{\log N}{N}$
random geometric	$\sim \log(N)$	$\sim \frac{\log N}{N}$

Solid lines are proportional to $N^{-1/2}$, $\log N/N$, $\log N/N$ and N^{-1}

Summary and Further Research

In the BGA, a **larger network** gives a **more accurate** averaging!

$$\text{Simulations suggest } \propto \frac{d_{\max}}{N} \quad \text{but} \quad \text{we have proved } \propto \frac{d_{\max}^2}{N\lambda_1}$$

Future research

- Close this gap!

Related known result: for 1- and 2-dimensional tori, $\lim_{N \rightarrow \infty} \beta(\infty) = 0$

F. Fagnani and P. Frasca. Broadcast gossip averaging: interference and unbiasedness in large Abelian Cayley networks. *IEEE Journal of Selected Topics in Signal Processing*, 5(4):866–875, 2011

- Extend this analysis to other randomized averaging algorithms

Further related reading

About the BGA and its performance:

F. Fagnani and S. Zampieri. Randomized consensus algorithms over large scale networks. *IEEE Journal on Selected Areas in Communications*, 26(4):634–649, 2008

T. C. Aysal, M. E. Yildiz, A. D. Sarwate, and A. Scaglione. Broadcast gossip algorithms for consensus. *IEEE Transactions on Signal Processing*, 57(7):2748–2761, 2009

T. C. Aysal, A. D. Sarwate, and A. G. Dimakis. Reaching consensus in wireless networks with probabilistic broadcast. In *Allerton Conf. on Communications, Control and Computing*, pages 732–739, Monticello, IL, September 2009

A. Tahbaz-Salehi and A. Jadbabaie. Consensus over ergodic stationary graph processes. *IEEE Transactions on Automatic Control*, 55(1):225–230, 2010